

# Towards Intelligent Control via Genetic Programming

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## Background and Aim of the work

### Aim:

- to investigate the use of Genetic Programming for the design of (intelligent) control systems

### Rationale:

- Challenges presented by Hypersonic Vehicles, Space Access Vehicles and Space Exploration require a new approach to control design to improve performances and robustness beyond the limits of current standard approaches.



An Intelligent Control (IC) application using GP on a modified version of the Goddard Rocket test case is proposed.

[1] B. Xu, Z. Shi. An overview on flight dynamics and control approaches for hypersonic vehicles. Science China Information Sciences. 2015

[2] Y. Xie, H. Huang, Y. Hu, G. Zhang. Applications of advanced control methods in spacecrafts: progress, challenges, and future prospects. Frontiers of Information Technology & Electronic Engineering. 2016.

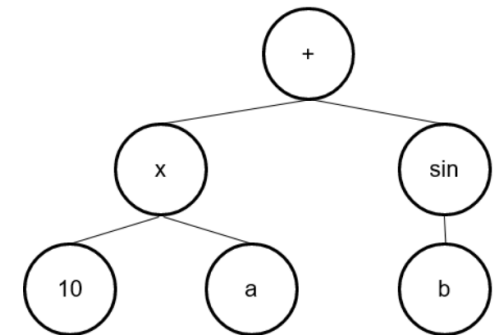
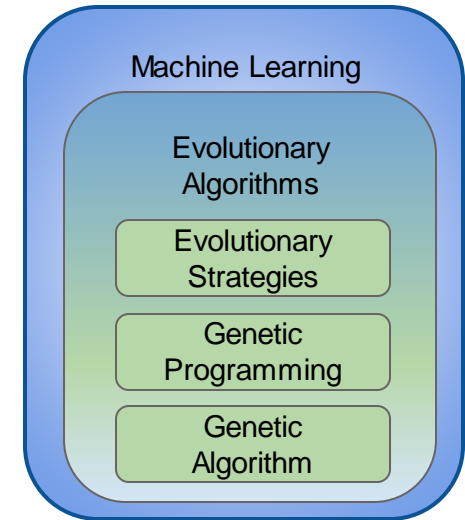
[3] D. Izzo, M. Martens, B. Pan. A survey on artificial intelligence trends in spacecrafts guidance dynamics and control. Astrodynamics. 2019

# Genetic Programming

- Originally introduced by Koza\* .
- Pertain to the class of Evolutionary Algorithms.
- Allow the creation of computer programs following the laws of natural evolution.

## GP applied to controller design:

- Both topology and parameters of the controller can be defined and optimized by GP\*\*.
- When nonlinearities are introduced in the system, the definition of a new kind of controller is necessary.
- Unlike other ML techniques, GP produces mathematical equations which could be analysed by the user with classical control techniques.



$$(10 \times a) + \sin(b)$$

\* J.R. Koza. Genetic Programming: On the Programming of Computers by Means of Natural Selection. 1992

\*\* J.R.Koza, M. A. Keane, J. Yu, F. H. Bennet III, W. Mydlowec. Automatic Creation of Human-Competitive Programs and Controllers by Means of Genetic Programming. Genetic Programming and Evolvable Machines. 2000

## What is intelligent control?

“An intelligent control system is designed so that it can autonomously achieve a high level goal, while its **components, control goals, plant models and control laws are not completely defined**, either because they were not known at the design time or because they changed unexpectedly.”\*



## Chosen approach to Intelligent control through GP:

- Thrust (control parameter) composed by the reference (obtained from open-loop optimal control) and a function of the relative errors created through GP.
- $f_{GP}$  is created when a change in the environment or in the system occurs.

$$T = T_{ref} + f_{GP}(e_1, \dots, e_n)$$

### Main issues:

- GP is computationally expensive.
- GP evaluation initial conditions must be defined a priori --> define evaluation time a priori!

\* P. Antsaklis. Defining Intelligent Control. Report of the Task Force on Intelligent Control, IEEE Control System Magazine, June 1994.

[7] G. N. Saridis. Toward the realization of Intelligent Controls. Proceedings of the IEEE, Vol. 67, No. 8. 1979

[8] C. Wilson, F. Marchetti, M. Di Carlo, A. Riccardi, E. Minisci. Intelligent Control: A Taxonomy. 8th International Conference on Systems and Control (ICSC). 2019.

# Test Case: Goddard Rocket 2 Controls

The chosen test case is a modified version of the Goddard Rocket as a first step for a future application on a more complex model of a space access vehicle. The optimal trajectory was obtained with a [Pseudospectral Collocation](#) method\*.

$$\begin{cases} \dot{r} = v_r \\ \dot{\theta} = \frac{v_t}{r} \\ \dot{v}_r = \frac{T_r}{m} - \frac{D_r}{m} - g + \frac{v_t^2}{r} \\ \dot{v}_t = \frac{T_t}{m} - \frac{D_t}{m} + \frac{v_t v_r}{r} \\ \dot{m} = -\frac{\sqrt{T_r^2 + T_t^2}}{g_0 I_{sp}} \end{cases}$$

$$D_r = \frac{1}{2} \rho v_r \sqrt{v_r^2 + v_t^2} C_d S$$

$$D_t = \frac{1}{2} \rho v_t \sqrt{v_r^2 + v_t^2} C_d S$$

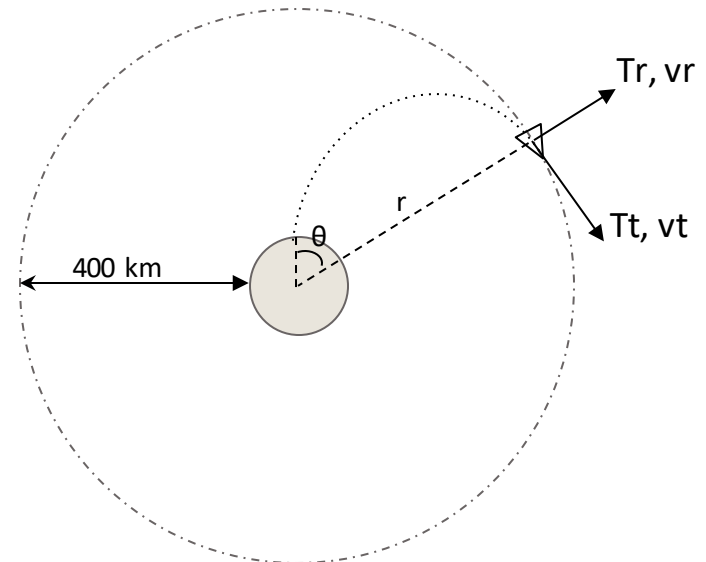
$$\rho = \rho_0 e^{-\beta r}$$

Where:

- $C_d = 0.6$
- $S = 4.0 \text{ m}^2$
- $I_{sp} = 300 \text{ s}$
- $m_0 = 100000 \text{ kg}$
- $m_p = 0.99 * m_0$

Mission goals:

- reach an altitude of 400 km
- using less propellant as possible

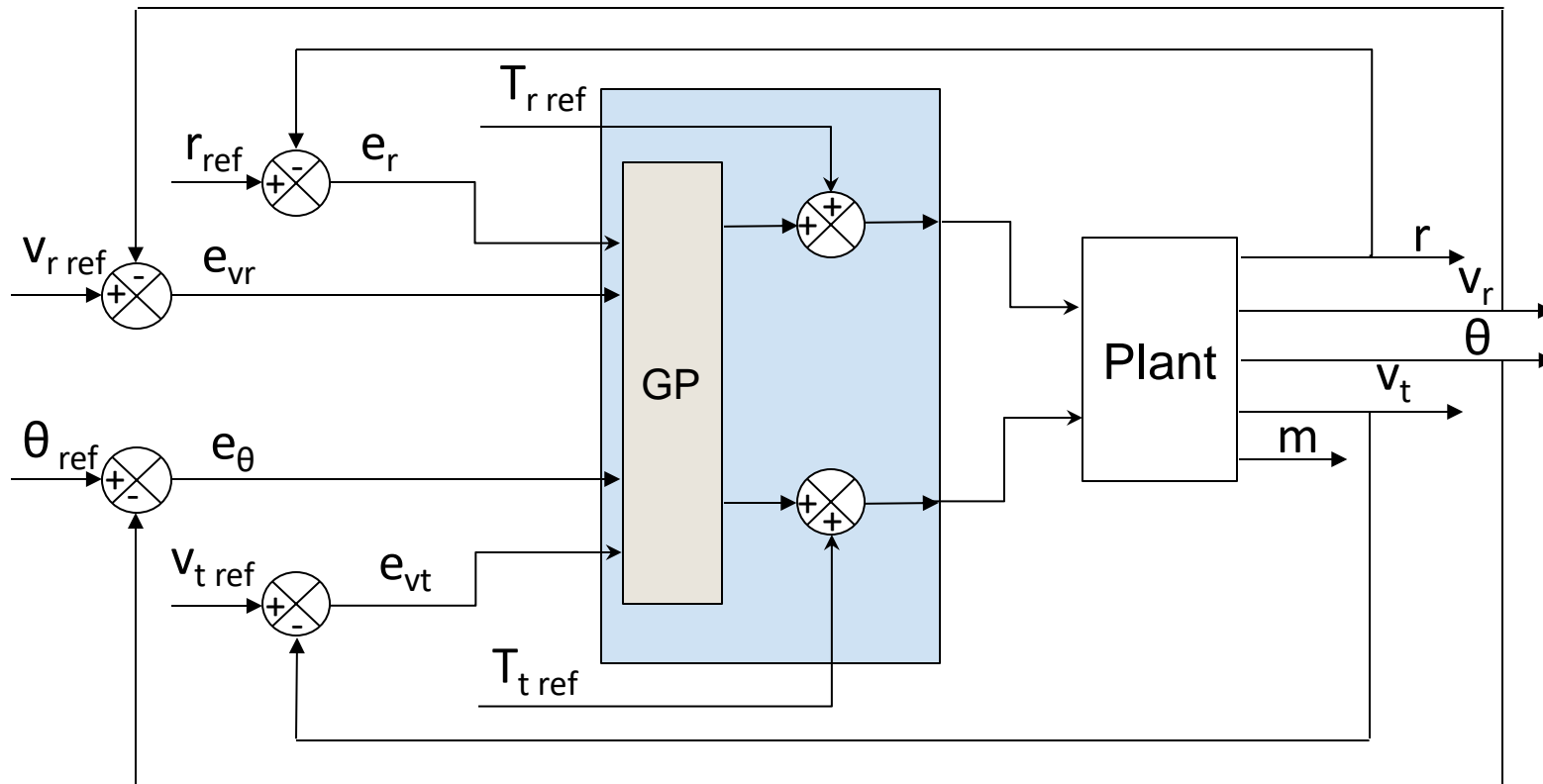


\* Istellartech. OpenGoddard open source library. <https://github.com/istellartech/OpenGoddard>. 2017

## Control scheme

$$T_r = T_{r_{ref}} + f_{GP_r}(e_r, e_{v_r})$$

$$T_t = T_{t_{ref}} + f_{GP_t}(e_\theta, e_{v_t})$$



## 3 scenarios are presented:

1. Variation of  $C_d$  at a random time and by a random magnitude.
2. Wind gust acting in a random altitude range with a constant random intensity in horizontal direction.

$$\begin{aligned}v_r &= v_r - v_{wind} \sin \theta \\v_t &= v_t - v_{wind} \cos \theta\end{aligned}$$

3. Real density model unknown at design time. Optimal trajectory obtained with simplified model and the real model is estimated during flight along with the controller law.

$$\rho = \rho_0 e^{-\beta r} \quad \longrightarrow \quad \rho = \rho_{ISA}(r)$$

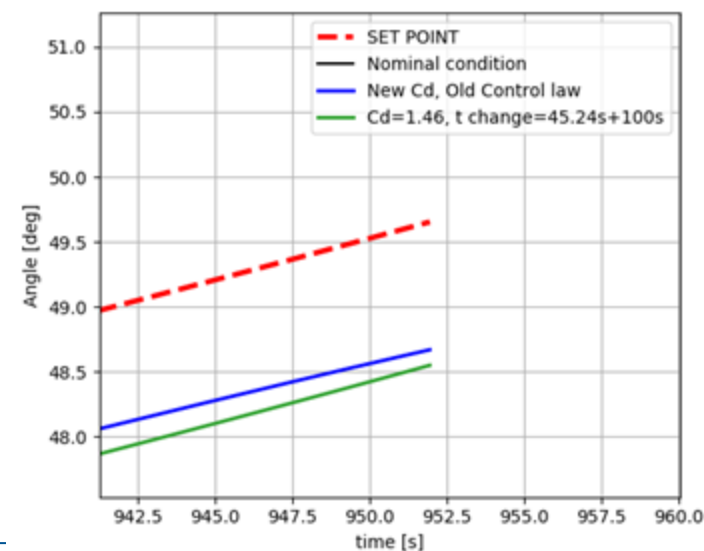
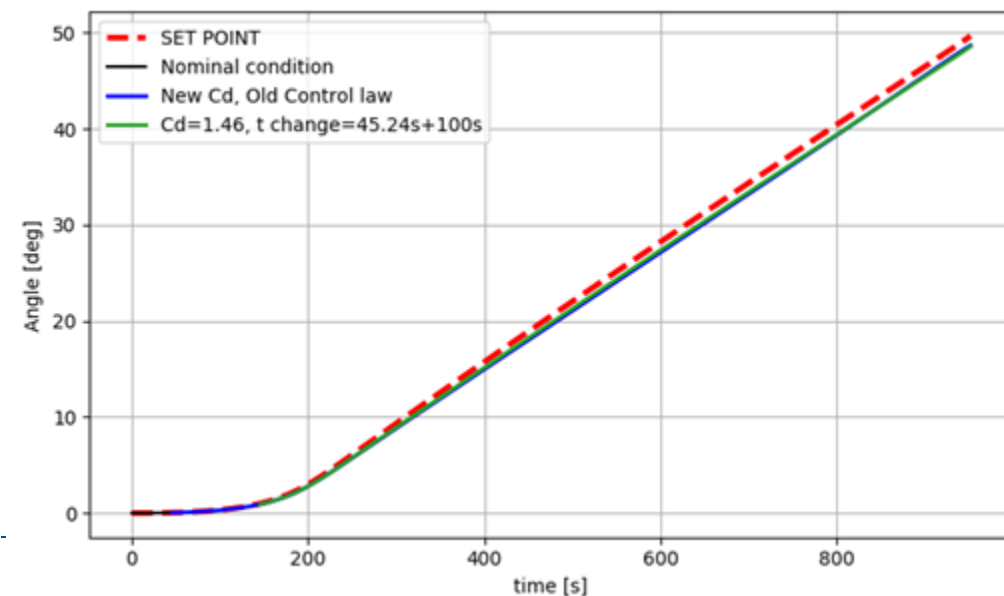
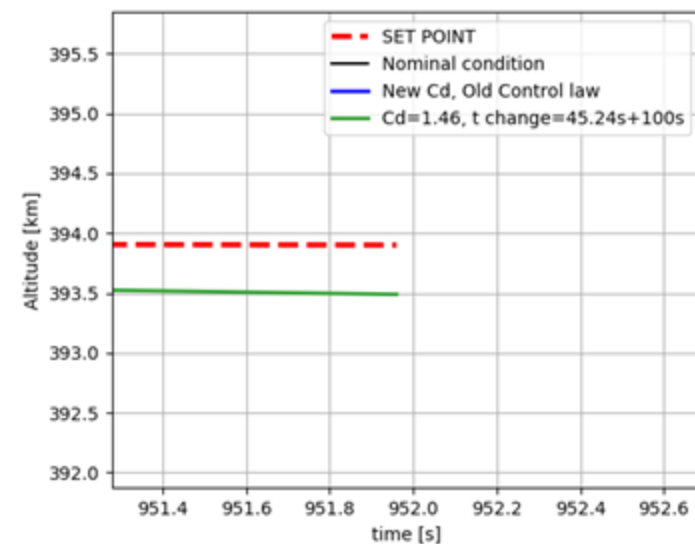
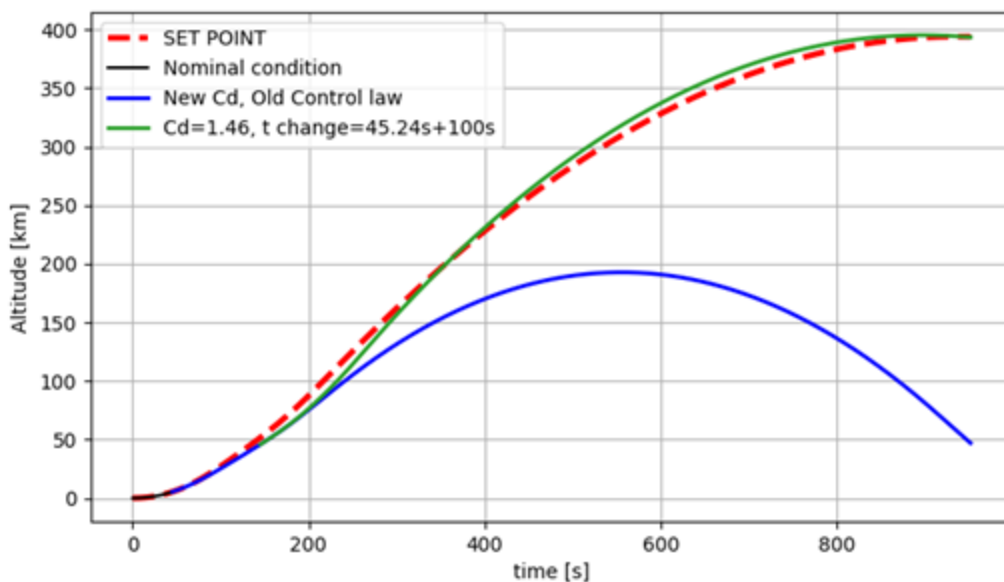
All simulations were run on a PC with 8GB of RAM and an Intel®Core™ i7-6700 CPU @3.40 GHz x 8 processors.

The scripts were coded in Python relying on the open source library DEAP\*.

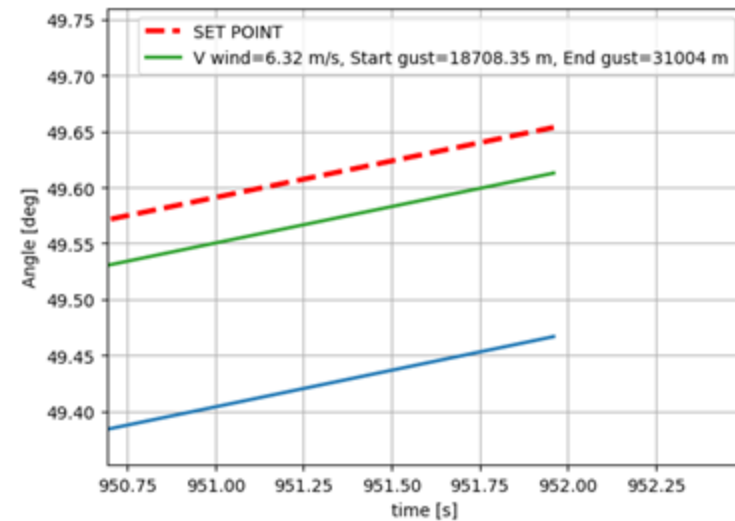
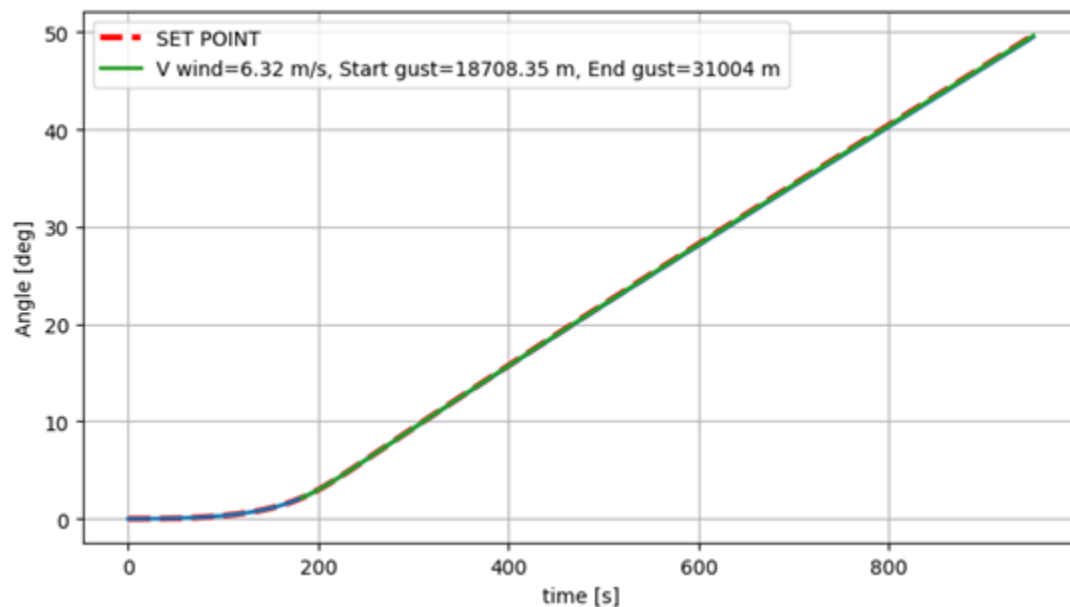
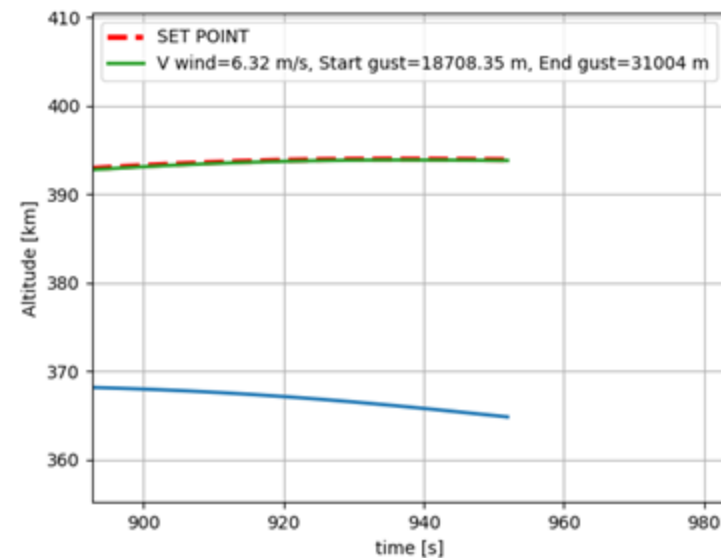
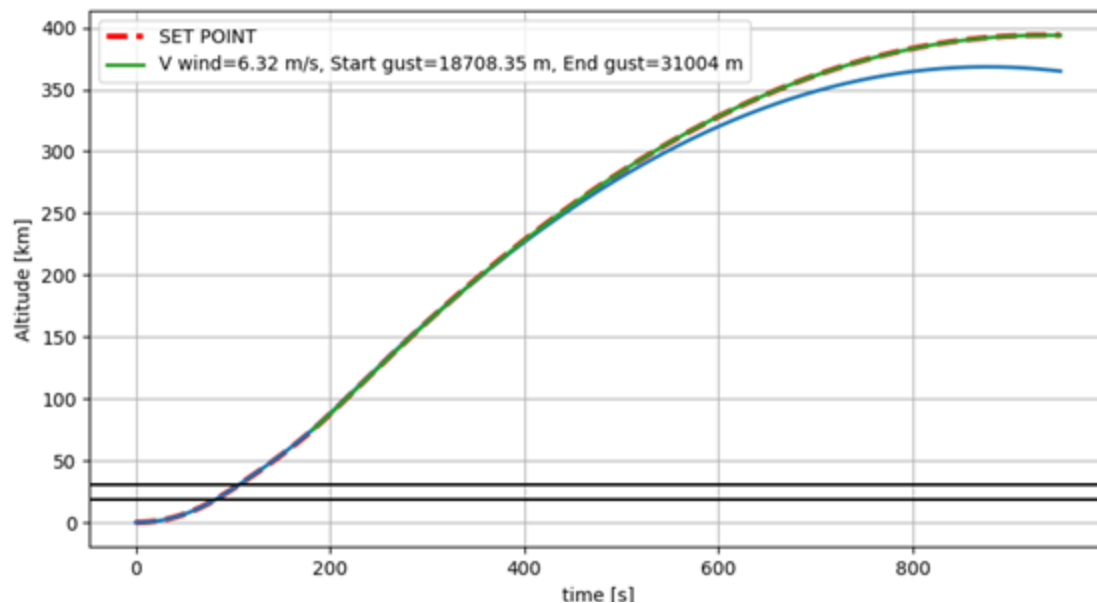
\* F.-A. Fortin and F.-M. De Rainville and M.-A. Gardner and M. Parizeau and C. Gagné. DEAP: Evolutionary Algorithms Made Easy. Journal of Machine Learning Research. 2012



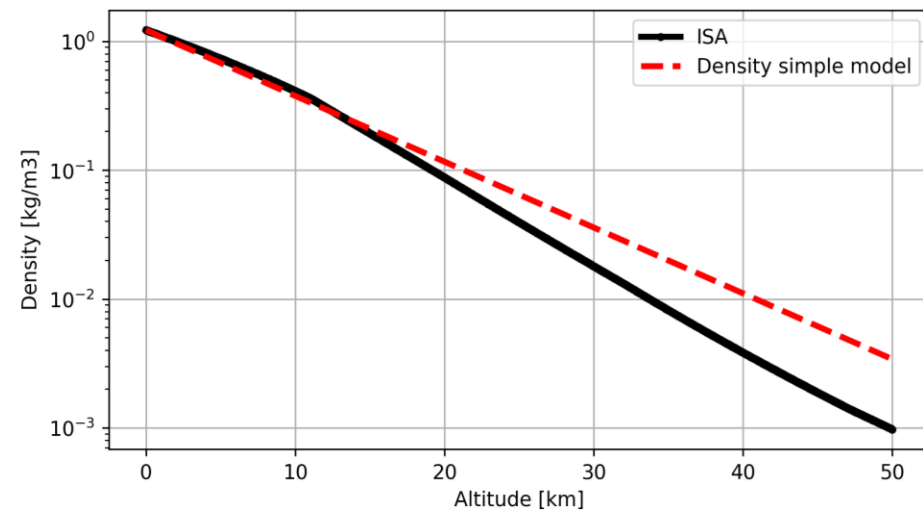
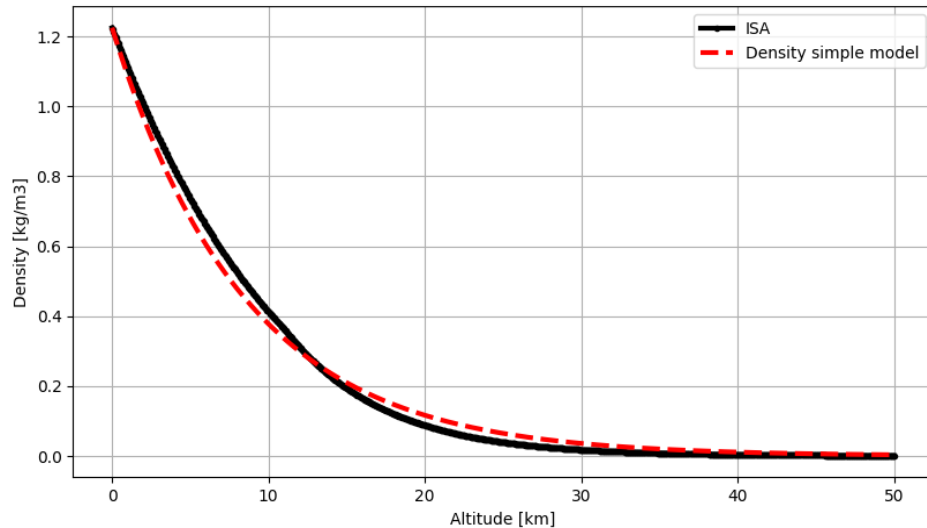
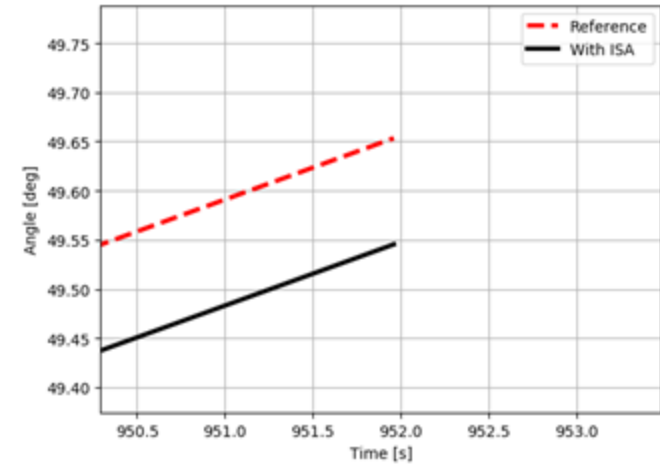
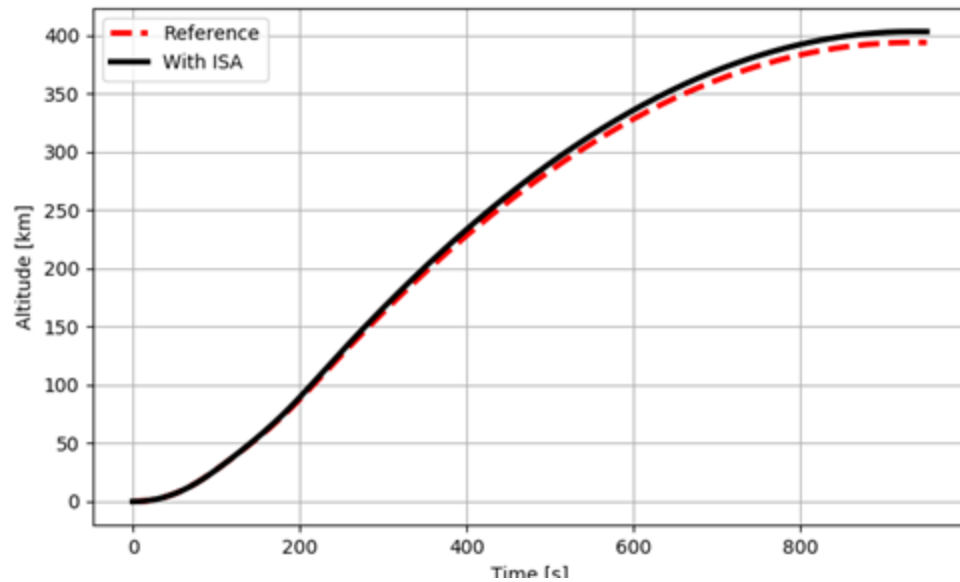
# Results: Scenario 1 - Cd variation



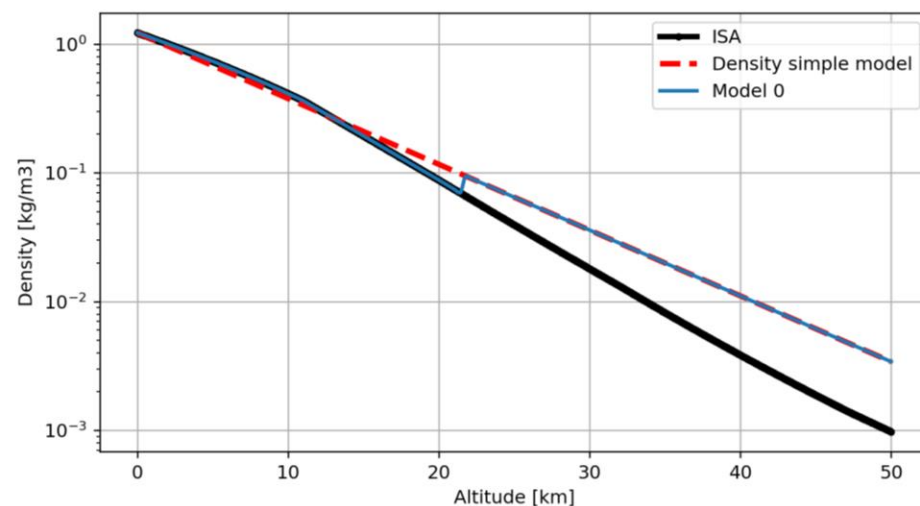
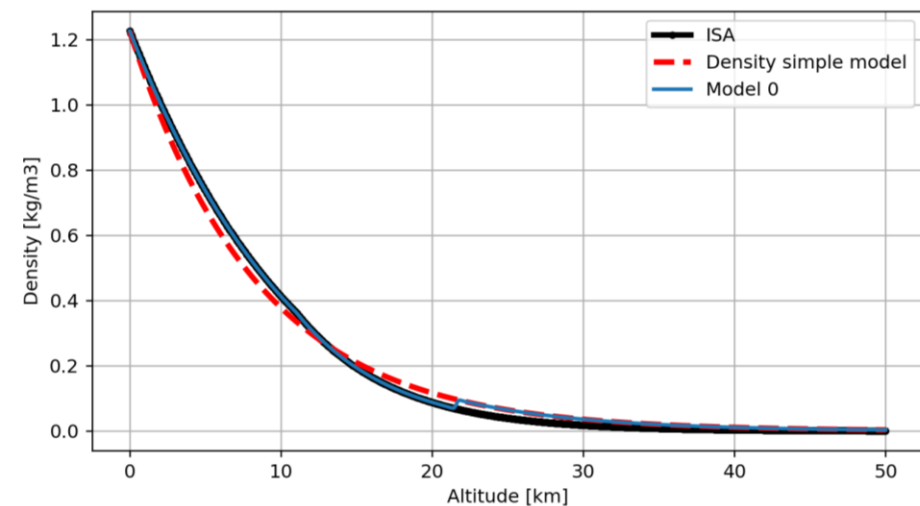
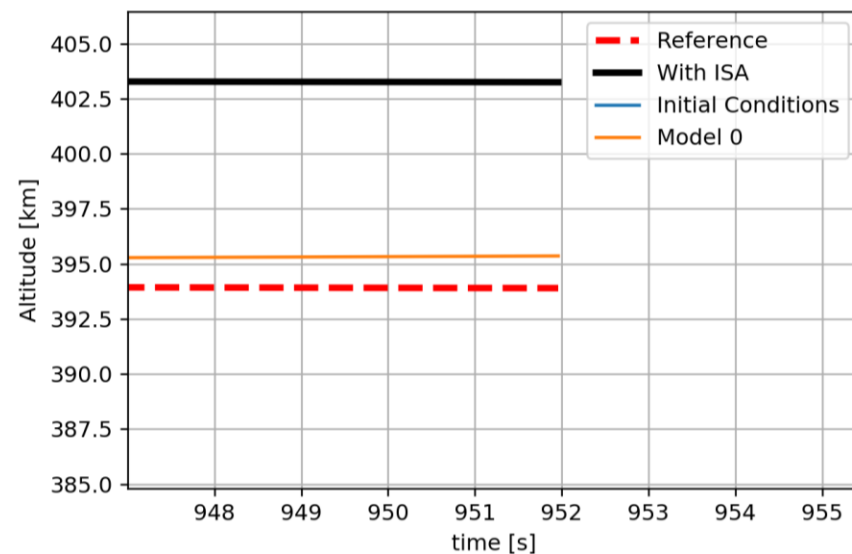
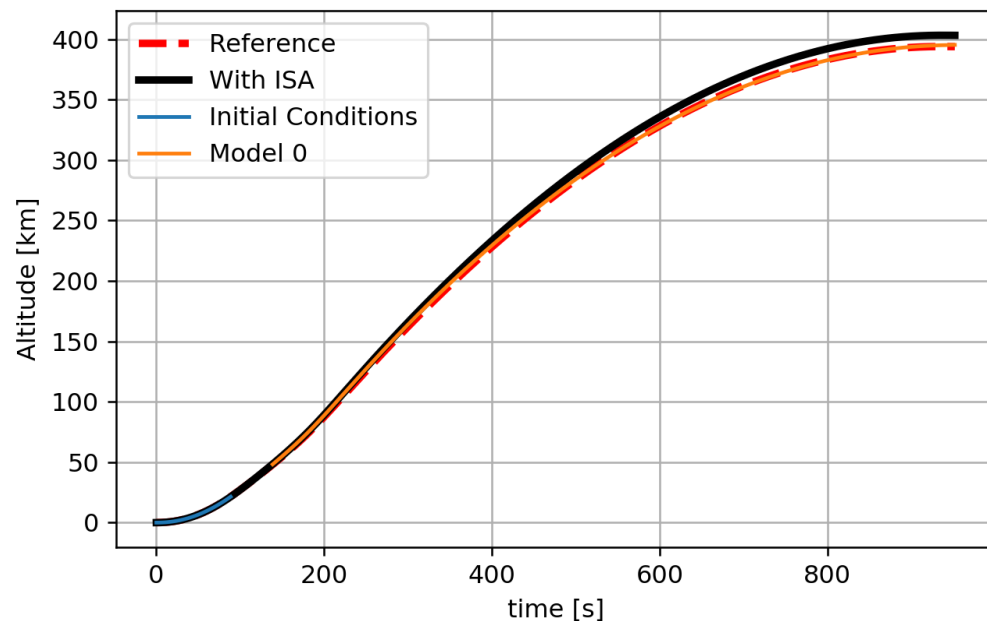
## Results: Scenario 2 - Wind Gusts



# Results: Scenario 3 - Unknown density model



# Results: Scenario 3 - Unknown density model



## Results: Statistics

	Cd Variation		Wind Gust		Density Model	
	No learning	Learning	No learning	Learning	No learning	Learning
GP Evaluations	100	100	100	100	270	199
Fixed evaluation time interval	100 s	100 s	100 s	100 s	50 s	50 s
Min evaluation time	3.63 s	8.11 s	8.76s	9.98 s	2.94 s	7.25 s
Max evaluation time	06h08m50s	01h27m37s	38m41s	55m59s	23m09s	18m41s
Time Constraint Success Rate	66%	76%	44%	62%	40%	49%
Range Constraint Success Rate	61%	64%	100%	100%	71%	82%
Total Success Rate	51%	60%	44%	62%	7%	46%

## Conclusions:

- GP can be a powerful tool to design a control law, with very good generalization capabilities.
- The biggest issues for a practical use of GP for IC is its computational cost.

## Future work:

- Implement more sophisticated methods to integrate models with "measured" data.
- Couple GP with an ANN, so to avoid the re-evaluation of the entire control law and perform the optimization of its components when needed.

## Summary:

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- [2] Y. Xie, H. Huang, Y. Hu, G. Zhang. Applications of advanced control methods in spacecrafts: progress, challenges, and future prospects. Frontiers of Information Technology & Electronic Engineering. 2016.
- [3] D. Izzo, M. Martens, B. Pan. A survey on artificial intelligence trends in spacecrafts guidance dynamics and control. Astrodynamics. 2019
- [4] J.R. Koza. Genetic Programming: On the Programming of Computers by Means of Natural Selection. 1992
- [5] J.R. Koza, M. A. Keane, J. Yu, F. H. Bennet III, W. Mydlowec. Automatic Creation of Human-Competitive Programs and Controllers by Means of Genetic Programming. Genetic Programming and Evolvable Machines. 2000
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- [9] Istellartech. OpenGoddard open source library. <https://github.com/istellartech/OpenGoddard>. 2017.
- [10] F.-A. Fortin and F.-M. De Rainville and M.-A. Gardner and M. Parizeau and C. Gagné. DEAP: Evolutionary Algorithms Made Easy. Journal of Machine Learning Research. 2012.



# *THANKS*

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